

# TOPICS COVERED

(NB: This list is not intended to be a complete list...)

## Solving linear systems

↳ Gauss-Jordan Elimination

↳ Row reduction and RREF matrices

## Geometry and linear systems

↳ dot product / angle formula.

## Matrices and Matrix operations

↳ addition, scalar multiplication, matrix product, transpose

## Vector spaces

↳ subspaces and subspace test

↳ span and linear independence

↳ \* Bases and dimension

$S \subseteq V$  n/  
 $S \neq \emptyset$  and  $ax + by \in S$   
for all scalar  $a, b$  and  
all  $x, y \in S$ .  $\Rightarrow S \leq V$

$S \subseteq V$  is lin. ind. when  
 $\sum c_i s_i = 0 \Rightarrow c_i = 0$  for all  $i$

## Linear maps

↳ linearity condition

↳ kernel and range spaces

↳ injectivity and surjectivity.

↳ Matrix representation

↳ Rank-Nullity Theorem  $\rightarrow \text{rank}(L) + \text{nullity}(L) = \dim(\text{dom}(L))$

↳ Linear operators \*  $L: V \rightarrow V$

$L$  is inj iff  $\ker(L) = \{0\}$

( $\approx$  null and column spaces)

## More on Matrices

↳ determinant

↳ elementary matrices \*

↳ inversion of matrices

## \* Change of Basis

### Eigenspaces

- ↳ Characteristic polynomial
- ↳ eigenvalues and eigenvectors
- ↳ Complex vector spaces

### Diagonalization of matrices / linear operators

- ↳ Similar matrices  $\rightsquigarrow B = PAP^{-1}$
- ↳ diagonalizability.  $\rightsquigarrow M = P \underset{\uparrow}{D} P^{-1}$

### Orthogonality (in $\mathbb{R}^n$ ).

- ↳ orthogonal projection
- \* ↳ orthogonal complement

$$\left[ \text{Col}(M)^\perp = \text{null}(M^T) \right]$$

- ↳ Gram-Schmidt process \*

- ↳ orthogonal matrices

$$\underline{A^{-1} = A^T} \quad (\text{i.e. } A^T A = I)$$

### Symmetric Matrices $\rightsquigarrow A^T = A$

- \* ↳ Transpose  $\rightsquigarrow (AB)^T = B^T A^T, (A+B)^T = A^T + B^T \dots$

- \* ↳ Real symmetric matrices have all eigenvalues real.

- ↳ Orthogonal diagonalizability

$$M = Q D Q^T$$

for  $Q$  orthogonal,  $D$  diagonal.

$$\rightsquigarrow \boxed{\begin{array}{l} M \text{ symmetric iff} \\ M \text{ ortho. diag'able.} \end{array}}$$

$$\ker(L) = \text{kernel of a linear map.}$$

$$= \{v \in \text{dom}(L) : L(v) = 0\}$$


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$$\text{null}(M) = \text{null space of matrix } M$$

$$= \text{solution set to } M\vec{x} = 0$$

$$= \ker(L_M) \quad \text{where } \text{Rep}_{\mathcal{E}_n, \mathcal{E}_m}(L_M) = M.$$

Point: kernel is associated to a linear map, whereas null space is associated to a matrix.

↳ often to compute a kernel of a linear map, we first compute the null space of an associated matrix, and then we convert that back into a kernel.

Ex: The linear map  $L: \mathcal{P}_3(\mathbb{R}) \rightarrow \mathbb{R}^3$  given by

$$L(a_0 + a_1x + a_2x^2 + a_3x^3) = \begin{pmatrix} a_0 + a_1 \\ a_1 + a_2 \\ a_0 + a_3 \end{pmatrix}.$$

to compute  $\ker(L)$ , we will compute null space of an associated matrix. Let  $B = \{1, x, x^2, x^3\} \subseteq \mathcal{P}_3(\mathbb{R})$ .

w.r.t.  $B$ ,  $L$  is represented by:

$$\left[ [L(1)]_{\mathcal{E}_3} \mid [L(x)]_{\mathcal{E}_3} \mid [L(x^2)]_{\mathcal{E}_3} \mid [L(x^3)]_{\mathcal{E}_3} \right]$$

$$= \begin{bmatrix} \overset{a_0}{1} & \overset{a_1}{1} & \overset{a_2}{0} & \overset{a_3}{0} \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 \end{bmatrix} = M.$$

$$\text{null}(M) = \text{null} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 \end{bmatrix} = \text{null} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & -1 & 0 & 1 \end{bmatrix} = \text{null} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$= \text{null} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \rightsquigarrow \begin{cases} a_0 = 0 \\ a_1 = 0 \\ a_2 = t \\ a_3 = 0 \end{cases} \leftarrow$$

$$\therefore \ker(L) = \{ v \in \mathcal{P}_3(\mathbb{R}) : a_0 + a_1x + a_2x^2 + a_3x^3 = v \}$$

$$a_0 = 0, a_1 = 0, a_2 = t \in \mathbb{R}, a_3 = 0$$

indicates

$$= \{ tx^2 : t \in \mathbb{R} \} \subseteq \mathcal{P}_3(\mathbb{R}). \leftarrow$$

by contrast:  $\text{null} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = \left\{ \begin{bmatrix} 0 \\ 0 \\ t \\ 0 \end{bmatrix} \in \mathbb{R}^4 : t \in \mathbb{R} \right\} \subseteq \mathbb{R}^4$

$M$  an  $m \times n$  matrix can represent a linear map  
from an  $n$ -dimensional space to an  $m$ -dimensional space.

Ex: Let  $M = \begin{bmatrix} 2 & 2 & 4 \\ 1 & 0 & 1 \\ 3 & 0 & 2 \end{bmatrix}$ .

$\hookrightarrow$  Rank:  $\text{null}(M) = \ker(L_M)$  where  $L_M: \mathbb{R}^n \rightarrow \mathbb{R}^m$   
 $\hookrightarrow L_M(x) = Mx$ .  
 and  $\text{col}(M) = \text{ran}(L_M)$  (same  $L_M$  as above)

Sol: To compute these, we compute  $\text{RREF}(M)$  because:

$$\text{null}(M) = \text{null}(\text{RREF}(M)) \quad \text{AND}$$

$\text{col}(M)$  has basis the columns of  $M$  corresp to leading 1's in  $\text{RREF}(M)$ .

$$\begin{array}{c} \downarrow \quad \downarrow \downarrow \\ \begin{bmatrix} 2 & 2 & 4 \\ 1 & 0 & 1 \\ 3 & 0 & 2 \end{bmatrix} \rightsquigarrow \begin{bmatrix} 1 & 0 & 1 \\ 2 & 2 & 4 \\ 3 & 0 & 2 \end{bmatrix} \rightsquigarrow \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 2 \\ 0 & 0 & -1 \end{bmatrix} \rightsquigarrow \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} \rightsquigarrow \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \end{array}$$

$\therefore \text{null}(M) = \text{null}(I_3)$  has  $\begin{cases} x = 0 \\ y = 0 \\ z = 0 \end{cases}$  so  $\text{null}(M) = \left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right\}$

has basis  $\emptyset$   $\text{col}(M)$  has basis  $\left\{ \begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 3 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 1 \\ 2 \end{bmatrix} \right\}$ .

→ Can't be simplified... row operations change column spaces...

Ex:  $\begin{bmatrix} 1 & 0 & 1 \\ -1 & 5 & -2 \\ 0 & 5 & -1 \end{bmatrix} \leftarrow$

Rank:

$$\text{nullity}(L_M) = 0 = \dim(\text{null}(M))$$

$$\text{rank}(L_M) = 3 = \dim(\text{col}(M))$$

$$\text{so } \dim(\mathbb{R}^3) = 3 = 0 + 3 = \text{nullity}(L_M) + \text{rank}(L_M)$$

→  $\text{nullity}(L_M) = 1$ ,  $\text{rank}(L_M) = 2$ . (check...)

$L$  is injective when for all  $x, y \in \text{dom}(L)$

we have  $L(x) = L(y)$  implies  $x = y$ . \*

→ "distinct inputs map to distinct outputs"

→  $L: V \rightarrow W$  is injective if and only if  $\ker(L) = 0$ .  
linear.

$L$  is surjective when for all  $y \in \text{col}(L)$  there is an  $x \in \text{dom}(L)$   
such that  $L(x) = y$ .

→ "every element of the codomain is an output".

→ Rank-Nullity Thm:  $\text{rank}(L) + \text{nullity}(L) = \dim(\text{dom}(L))$ .

if  $\text{rank}(L) = \dim(\text{col}(L))$ , then  $L$  is surjective.

$L$  is bijective when it is both surjective and injective.

→ Linear  $L$  is bijective iff  $L$  is an isomorphism..